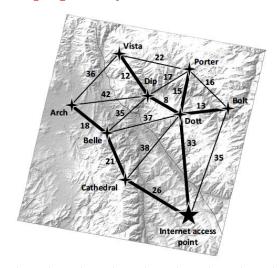


Lecture 14 Minimum Spanning Trees (MSTs): Prim, Kruskal

CS 161 Design and Analysis of Algorithms
Ioannis Panageas

Application: Connecting a Network

- Suppose the remote mountain country of Vectoria has been given a major grant to install a large Wi-Fi the center of each of its mountain villages.
- Communication cables can run from the main Internet access point to a village tower and cables can also run between pairs of towers.
- The challenge is to interconnect all the towers and the Internet access point as cheaply as possible.



Application: Connecting a Network

- We can model this problem using a graph, G, where each vertex in G is the location of a Wi-Fi the Internet access point, and an edge in G is a possible cable we could run between two such vertices.
- Each edge in G could then be given a weight that is equal to the cost of running the cable that that edge represents.
- Thus, we are interested in finding a connected acyclic subgraph of G that includes all the vertices of G and has minimum total cost.
- Using the language of graph theory, we are interested in finding a minimum spanning tree (MST) of G.

Minimum Spanning Trees

Spanning subgraph

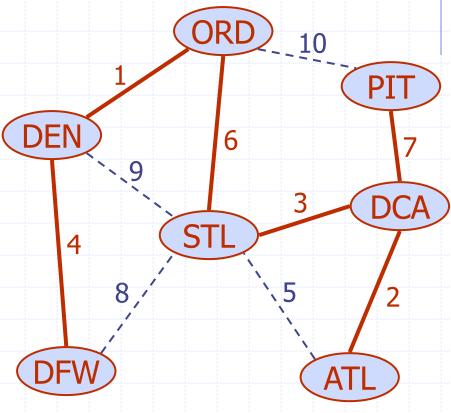
Subgraph of a graph G
 containing all the vertices of G

Spanning tree

Spanning subgraph that is itself a (free) tree

Minimum spanning tree (MST)

- Spanning tree of a weighted graph with minimum total edge weight
- Applications
 - Communications networks
 - Transportation networks



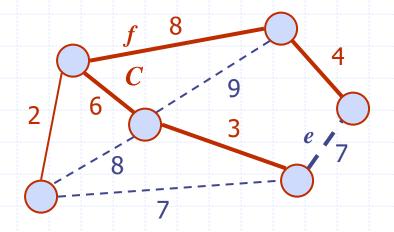
Cycle Property

Cycle Property:

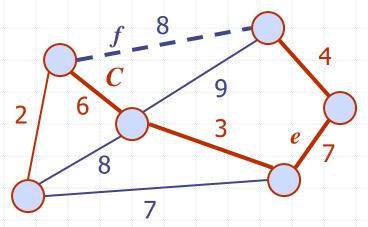
- Let T be a minimum spanning tree of a weighted graph G
- Let e be an edge of G that is not in T and C let be the cycle formed by e with T
- For every edge f of C, weight(f) ≤ weight(e)

Proof:

- By contradiction
- If weight(f) > weight(e) we can get a spanning tree of smaller weight by replacing e with f



Replacing f with e yields a better spanning tree



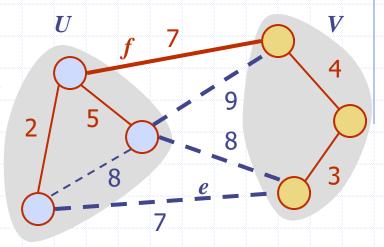
Partition Property

Partition Property:

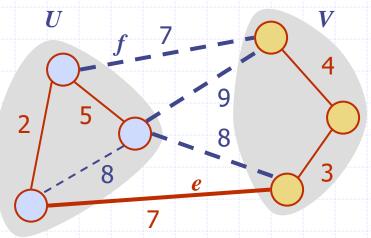
- Consider a partition of the vertices of G into subsets U and V
- Let e be an edge of minimum weight across the partition
- There is a minimum spanning tree of G containing edge e

Proof:

- Let T be an MST of G
- If *T* does not contain *e*, consider the cycle *C* formed by *e* with *T* and let *f* be an edge of *C* across the partition
- By the cycle property, weight(f) ≤ weight(e)
- Thus, weight(f) = weight(e)
- We obtain another MST by replacing f with e



Replacing f with e yields another MST



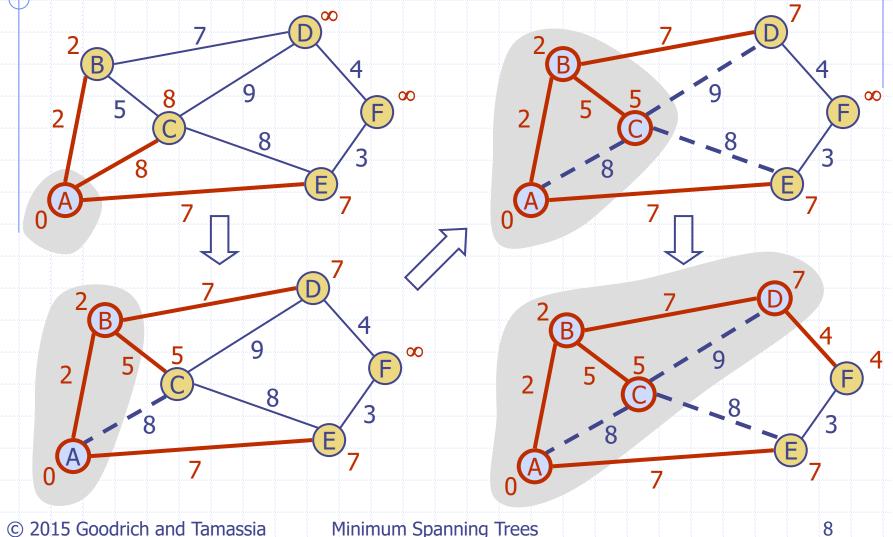
Prim-Jarnik's Algorithm

- Similar to Dijkstra's algorithm
- We pick an arbitrary vertex s and we grow the MST as a cloud of vertices, starting from s
- We store with each vertex v label d(v) representing the smallest weight of an edge connecting v to a vertex in the cloud
- At each step:
 - We add to the cloud the vertex u outside the cloud with the smallest distance label
 - lacktriangle We update the labels of the vertices adjacent to u

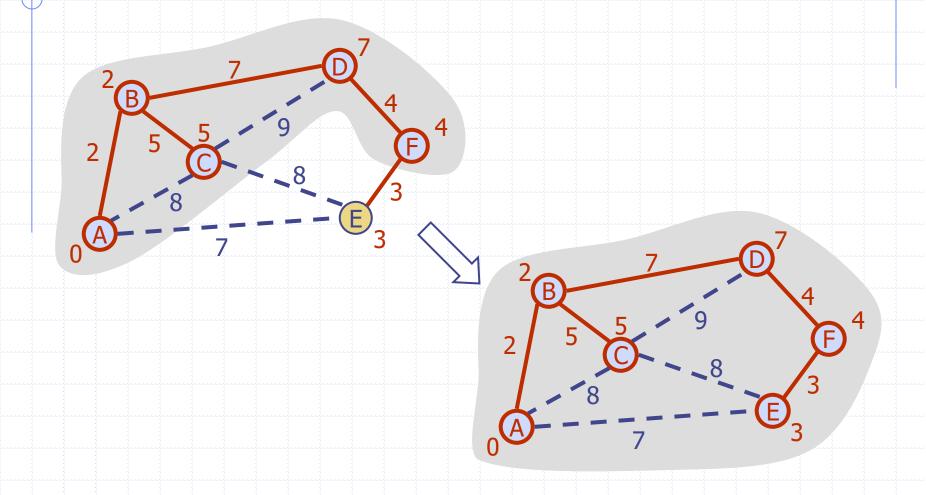
Prim-Jarnik Pseudo-code

```
Algorithm PrimJarníkMST(G):
   Input: A weighted connected graph G with n vertices and m edges
   Output: A minimum spanning tree T for G
    Pick any vertex v of G
    D[v] \leftarrow 0
    for each vertex u \neq v do
        D[u] \leftarrow +\infty
    Initialize T \leftarrow \emptyset.
    Initialize a priority queue Q with an item ((u, \text{null}), D[u]) for each vertex u,
    where (u, \text{null}) is the element and D[u] is the key.
    while Q is not empty do
         (u, e) \leftarrow Q.removeMin()
         Add vertex u and edge e to T.
         for each vertex z adjacent to u such that z is in Q do
             // perform the relaxation procedure on edge (u, z)
             if w((u,z)) < D[z] then
                  D[z] \leftarrow w((u,z))
                  Change to (z, (u, z)) the element of vertex z in Q.
                  Change to D[z] the key of vertex z in Q.
    return the tree T
```

Example



Example (contd.)



Analysis

- Graph operations
 - We cycle through the incident edges once for each vertex
- Label operations
 - We set/get the distance, parent and locator labels of vertex z $O(\deg(z))$ times
 - Setting/getting a label takes O(1) time
- Priority queue operations
 - Each vertex is inserted once into and removed once from the priority queue, where each insertion or removal takes $O(\log n)$ time
 - The key of a vertex w in the priority queue is modified at most deg(w) times, where each key change takes $O(\log n)$ time
- Prim-Jarnik's algorithm runs in $O((n+m)\log n)$ time provided the graph is represented by the adjacency list structure
 - Recall that $\sum_{v} \deg(v) = 2m$
- \Box The running time is $O(m \log n)$ since the graph is connected

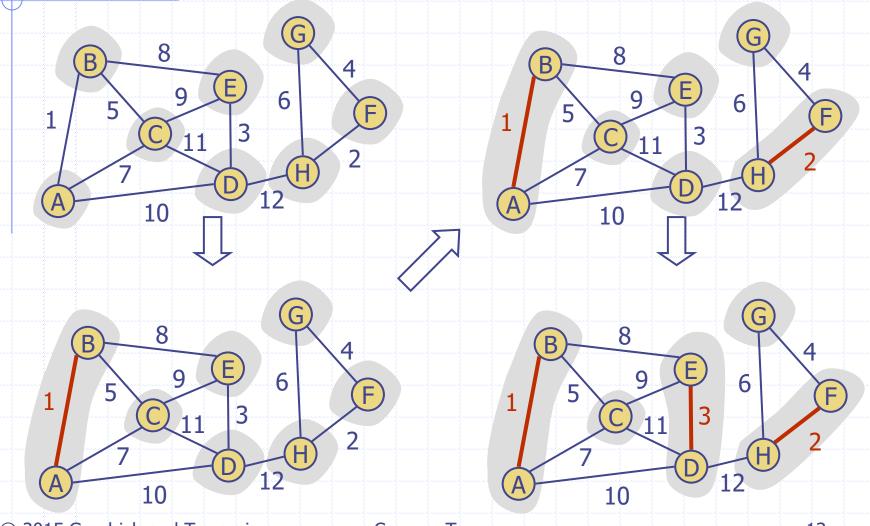
Kruskal's Approach

- Maintain a partition of the vertices into clusters
 - Initially, single-vertex clusters
 - Keep an MST for each cluster
 - Merge "closest" clusters and their MSTs
- A priority queue stores the edges outside clusters (or you could even sort the edges)
 - Key: weight
 - Element: edge
- At the end of the algorithm

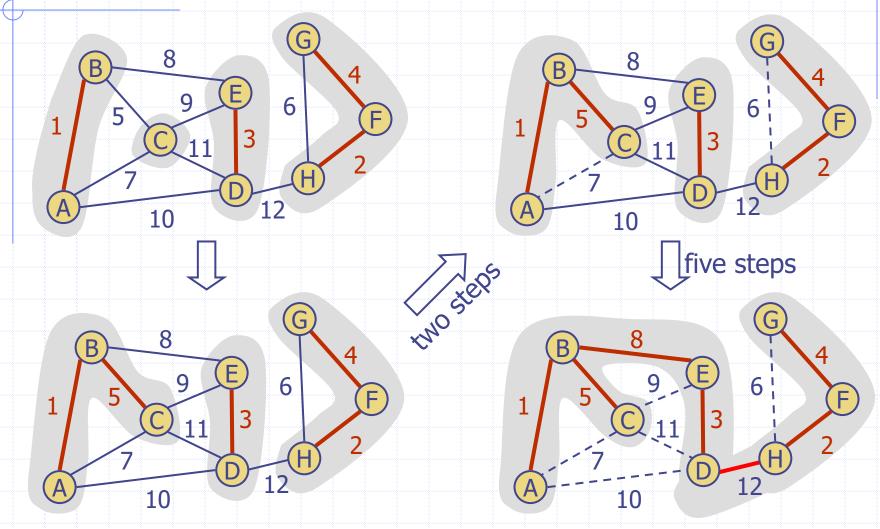
Kruskal's Algorithm

```
Algorithm KruskalMST(G):
   Input: A simple connected weighted graph G with n vertices and m edges
   Output: A minimum spanning tree T for G
    for each vertex v in G do
        Define an elementary cluster C(v) \leftarrow \{v\}.
    Let Q be a priority queue storing the edges in G, using edge weights as keys
                // T will ultimately contain the edges of the MST
    while T has fewer than n-1 edges do
        (u, v) \leftarrow Q.\mathsf{removeMin}()
        Let C(v) be the cluster containing v
        Let C(u) be the cluster containing u
        if C(v) \neq C(u) then
            Add edge (v, u) to T
             Merge C(v) and C(u) into one cluster, that is, union C(v) and C(u)
    return tree T
```

Example of Kruskal's Algorithm



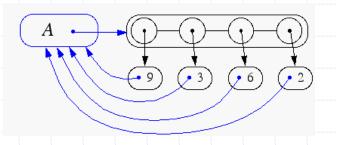
Example (contd.)



Data Structure for Kruskal's Algorithm

- The algorithm maintains a forest of trees
- A priority queue extracts the edges by increasing weight
- An edge is accepted it if connects distinct trees
- We need a data structure that maintains a partition, i.e., a collection of disjoint sets, with operations:
 - makeSet(u): create a set consisting of u
 - find(u): return the set storing u
 - union(A, B): replace sets A and B with their union

List-based Partition



- Each set is stored in a sequence
- Each element has a reference back to the set
 - operation find(u) takes O(1) time, and returns the set of which u is a member.
 - in operation union(A,B), we move the elements of the smaller set to the sequence of the larger set and update their references
 - the time for operation union(A,B) is min(|A|, |B|)
- Whenever an element is processed, it goes into a set of size at least double, hence each element is processed at most log n times

Partition-Based Implementation

- Partition-based version of Kruskal's Algorithm
 - Cluster merges as unions
 - Cluster locations as finds
- □ Running time $O((n + m) \log n)$
 - Priority Queue operations: $O(m \log n)$
 - Union-Find operations: $O(n \log n)$